

Performance Improvements in Temperature Reconstructions of 2-D Tunable Diode Laser Absorption Spectroscopy (TDLAS)

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Performance improvement was attained in data reconstructions of 2-dimensional tunable diode laser absorption spectroscopy (TDLAS). Multiplicative Algebraic Reconstruction Technique (MART) algorithm was adopted for data reconstruction. The data obtained in an experiment for the measurement of temperature and concentration fields of gas flows were used. The measurement theory is based upon the Beer-Lambert law, and the measurement system consists of a tunable laser, collimators, detectors, and an analyzer. Methane was used as a fuel for combustion with air in the Bunsen-type burner. The data used for the reconstruction are from the optical signals of 8-laser beams passed on a cross-section of the methane flame. The performances of MART algorithm in data reconstruction were validated and compared with those obtained by Algebraic Reconstruction Technique (ART) algorithm.

Keywords: Data Reconstruction, TDLAS, MART, ART.

Introduction

Temperature distribution on the cross-section of a combustion flame allows for exact analyses on the combustion phenomena. Measurements on the exhaust gases such as NO_x, SO_x, CO_x and CH_x from plants or factories allows for preventing air pollutions and global warming. TDLAS (tunable diode laser absorption spectroscopy) has been used for these purposes.

For the measurement on the concentrations of the gases, there have been several representative pointwise techniques. Wang et al. [1] used a WAV (surface acoustic wave) type sensor to detect CO₂ and humidity. Stemhagen et al.[2] used a WAV sensor to detect gas and temperature. Fine et al. [3] constructed a semiconductor-type sensor to detect CO, NO_x, NH₃ and CO₂ gases. Chen [4]

proposed a thermal conductivity-type sensor for detection of H₂ gas. Dossi et al.[5] adopted an electrochemical-type sensor for NH₄ gas. Since all of these conventional measurement techniques were based on pointwise measurements, the distribution of temperature or concentration can't be measured. As field-wise measurement techniques, Ko et al. [6] proposed a digital speckle method which allows for the measurements of spatial distribution of CO₂ concentration. Since gas concentration is affected by the gas temperature, temperature distribution should be measured to predict the gas concentrations. Deguchi et al. [7] proposed a technique which can quantify the concentration distributions of H₂O vapor by measuring the temperature distribution of a gas flame. They used a tunable laser and introduced 2-dimensional array of the laser beam, in which 8 laser beam lines were installed and the

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intensity of the laser lines were sensed by 8 detectors (photodiode). The intensity signals from 8 detectors were used to reconstruct the spectrum patterns of a gas. They used the ART (algebraic reconstruction technique) [8] method for reconstructing the concentration and temperature fields. In the meantime, Doh et al. [9] developed a tomographic PTV (particle tracking velocimetry) in which spatial distribution of velocity vectors are obtained from a set of tomography images that were reconstructed by the MART (multiplicative algebraic reconstruction technique) method. In this study, it was known that the MART algorithm allows less calculation load and less calculation errors than the ART algorithm. In this study, performance of the MART algorithm is shown by using the experimental data obtained by the Deguchi group [7].

Temperature vs Relative Intensity

Fig. 1 shows major spectroscopic steps, such as absorption, fluorescence and photoionization, from which spectra signals can be measured for the investigations of gas properties. The absorption spectroscopy is fundamentally based upon Beer-Lambert Law as shown in Fig. 2. When the incident laser beam passes through the uniform medium, the ratio of the transmitted light intensity (I_t) against the incident light intensity (I_0) is

$$I_t(\lambda) = I_0(\lambda) \cdot \exp\{-\alpha_\lambda\} \quad (1)$$

inversely proportional to the power of the absorbance coefficient (α_λ) Eq. (1). The intensity ratio of incident light and transmitted light depends on the mole fraction. The number density of the measured species n is related to the amount of light absorbed as in the following formula.

$$\begin{aligned} I_t(\lambda)/I_0(\lambda) &= \exp\{-A_\lambda\} \\ &= \exp\left\{\sum_i (n_i L \alpha_i)\right\} \\ &= \exp\left\{\sum_i \left(n_i L \sum_j S_{i,j}(T) G_{vi,j}\right)\right\} \end{aligned} \quad (2)$$

Here, I_t and I_0 are the transmitted and incident laser intensities and A_λ represents the spectral absorbance. n_i is the number density of species i , L is the optical path length, α_i is absorption coefficient, S_{ij} is the temperature dependent absorption linestrength of the absorption line j , and $G_{vi,j}$ is the line-broadening function. Voigt profile was used for the line-broadening function. Three absorption lines located at 1388.135nm(#1), 1388.326nm(#2), and 1388.454nm (#3) were selected to measure temperature and H₂O concentration. Fig. 1(a) and Fig. 1(b) show theoretical H₂O absorption spectra calculated using the HITRAN 2008 database [10] at 300K, 500K, 600K and 1000K. Fig. 3 shows the temperature dependences of the three absorption lines. In order to increase the linearity

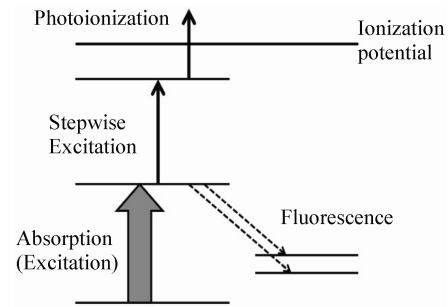
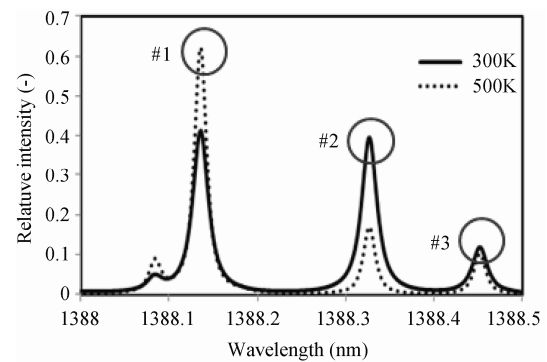
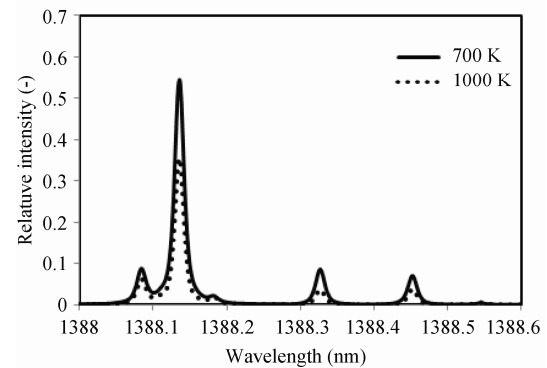


Fig. 1 Major spectroscopic steps of atoms



(a) 300K, 500K at 0.1Mpa



(b) 700K, 1000K at 0.1Mpa

Fig. 2 Relative intensity of theoretical H₂O absorption spectra (1388nm ~ 1388.6nm)

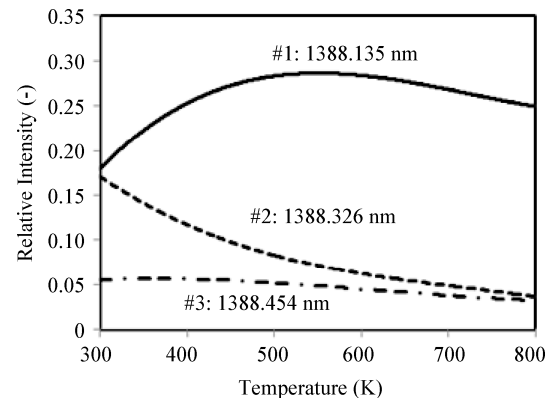


Fig. 3 Relative intensity dependencies on temperature variations for three absorption lines of H₂O.

Experiments and Temperature Reconstructions

Temperature (K)	#1/#2 Intensity Ratio (°)	#3/#2 Intensity Ratio (°)
300	1.0	0.4
400	2.0	0.5
500	3.0	0.6
600	4.0	0.7
700	5.0	0.8
800	7.0	0.9

The schematic diagram illustrates the experimental setup for the study of CH₄ oxidation. The system includes a Tunable laser connected to a Fiber splitter, which is linked to a Thermocouple and a Burner. A Flowmeter and Valve are connected to a CH₄ tank, which feeds into a Flowmeter. The Flowmeter is connected to a Compressor, which feeds into another Flowmeter. The Burner is positioned over a sample surface, and the Analyzer is connected to the system. The diagram also shows the flow of CH₄ (dashed line) and Air (solid line) through the system.

where the subscript i represent the grid point ($i = 1, 2, \dots$), λ is transition frequency. Because the integrated absorbance is dependent on both temperature and concentration, the temperature distribution has to be calculated by more than two different absorbance values. In this study, temperature and H_2O concentration were determined at each analysis grids using a multifunction minimization method to minimize the spectral fitting error at 1338.0–1338.6 nm. This minimization method was calculated by the following equation.

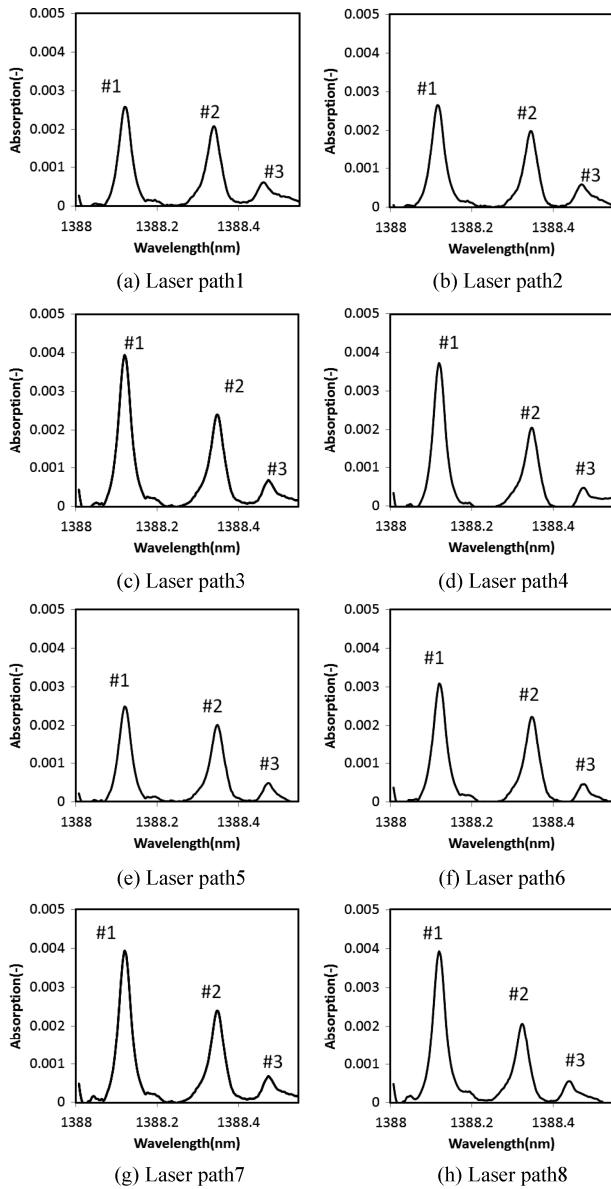


Fig. 7 Measured absorption spectra (H_2O) for 8-path lasers

$$Error = \sum \left\{ (A_{\lambda,i})_{theory} - (A_{\lambda,i})_{experiment} \right\}^2 \quad (4)$$

The measured H_2O absorption spectra was normalized and compared to those of theoretical spectra to minimize the MSE (Mean Squared Errors). The error was calculated and minimized from each of all laser paths. This procedure is explained in Fig. 8. All values of $\alpha_{v1,j}(i)$ at all grids points were iteratively calculated using the ART and MART methods until convergence. The final values of $T(i, j)$ and $n(i, j)$ after iterations were then reconstructed for temperature and concentration fields. For data reconstructions, two algorithms, ART and MART were used and their performances were compared.

Eq. (5) represents the equation used for ART (alge-

braic reconstruction technique) algorithm. The unknown values of absorption coefficients (α_{λ}) are calculated implicitly.

$$\alpha_{v1,j}(i)^{(k+1)} = \alpha_{v1,j}(i)^{(k)} + \beta \frac{A_{v1,j} - \sum_{i=1}^I \alpha_{v1,j}(i)^{(k)} \cdot L_{ij}}{\sum_{i=1}^I L_{ij}^2} \quad (5)$$

Eq. (6) represents the equation used for MART (multiplicative algebraic reconstruction technique) method.

$$\alpha_{v1,j}(i)^{(k+1)} = \alpha_{v1,j}(i)^{(k)} + \left[\frac{A_{v1,j}}{\sum_{i=1}^I \alpha_{v1,j}(i)^{(k)} \cdot L_{ij}} \right]^{\beta L_{ij}} \quad (6)$$

Here, $\alpha_{v1,j}(i)$ is expressed as Eq. (7).

$$\alpha_{v1,j}(i) = n(i, j) \cdot S_{i,j} \{T(i, j), v1\} \cdot G_{v1,j} \cdot P \quad (7)$$

With initial or previous values of $n(i, j)$ and $T(i, j)$, $\alpha_{v1,j}(i)$ values were iteratively calculated using Eq. (7).

Implicit solution was performed until all values of $\alpha_{v1,j}(i)$ at all grids points converged to constant values. k is the iteration number of MART algorithm, and β is a relaxation parameter. $\alpha_{v1,j}$ is absorption coefficient at the wavelength $v1$ (1388.135nm), $A_{v1,j}$ is the absorbance obtained from experiments.

Iterative calculation was made for #1($v1$, 1388.135nm) at first, next for #2($v2$, 1388.326nm). When performing iterative calculations, the temperatures measured by thermocouples were used as the initial values. Then, the initial values of gas concentrations were obtained by the initial values of temperature using Eq. (3). Absorption coefficients were newly calculated using the absorption values from the horizontal laser 1 to 4 using Eq. (5). After iterative calculations for the horizontal lasers (1~4), the absorption coefficients of the vertical lasers (5~8) were also renewed.

Fig. 9 shows the mean squared errors (MSE) obtained by ART algorithm for all iteration numbers. β values were changed in the range of ($0.05 < \beta < 1$). For the case

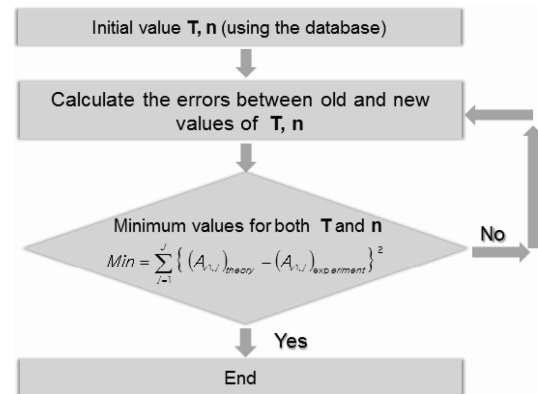


Fig. 8 Procedure for calculation of absorption coefficients

of ART algorithm, the MSE was the smallest values when β was set to 0.1. The MSE was saturated within the iteration number 50. This implies that the iteration numbers do not need to be large number if an optimal relaxation value (β) is adopted.

Fig. 10 shows the mean squared errors (MSE) obtained by MART algorithm for all iteration numbers. β values were also changed in the range of $(0.05 < \beta < 1)$. For the case of ART algorithm, the MSE was the smallest values when β was set to 0.5. The MSE was saturated within the iteration number 10. This implies that the calculation time is 5 times shorter than that of ART algorithm if an optimal relaxation value (β) is adopted.

Fig. 11 shows comparison of calculation performances between ART and MART algorithms. It can be inferred that the magnitudes of errors by MART algorithm are lower than those by ART algorithm with rapid calculation convergences.

Fig. 12 and Fig. 13 show the distributions of temperature and concentration obtained by MART algorithm. It can be seen that the peak region are biased to the edges. This is due to the fact that the center of flame burner (high temperature region) was installed at the edge loca-

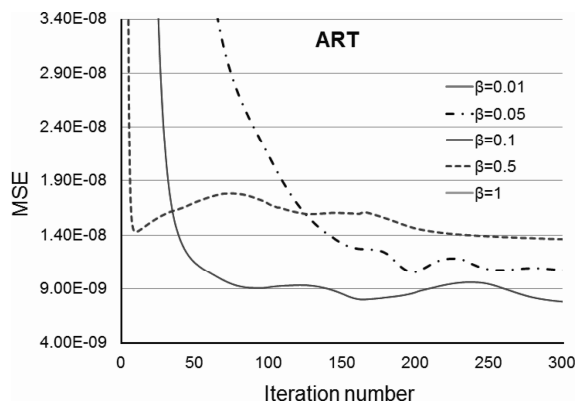


Fig. 9 Variations of mean square error (MSE) with iteration numbers for relaxation parameters in ART algorithm.

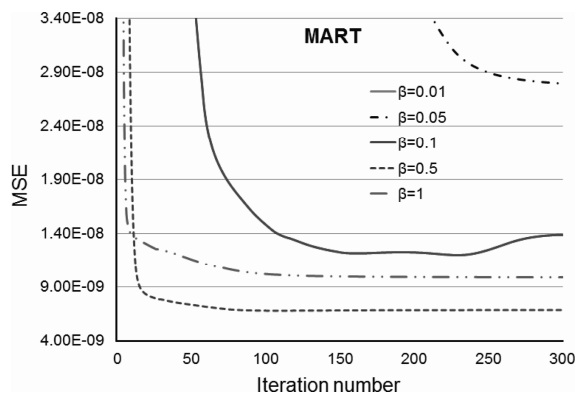


Fig. 10 Variations of mean square error (MSE) with iteration numbers for relaxation parameters in MART algorithm.

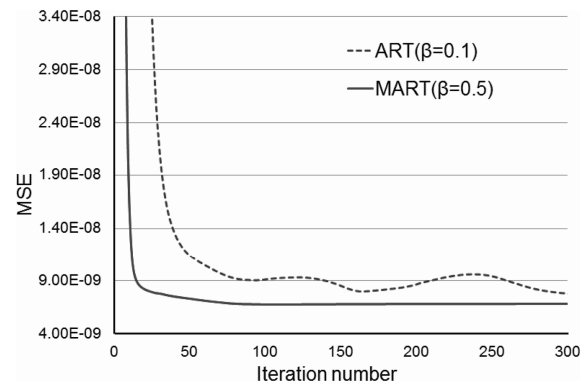


Fig. 11 Comparisons between ART algorithm and MART algorithm

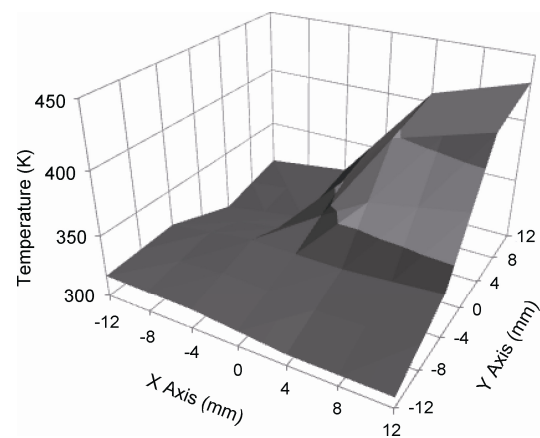


Fig. 12 Temperature distribution obtained by MART

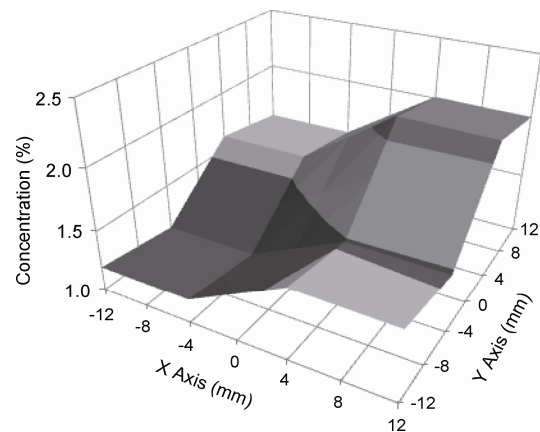


Fig. 13 Concentration distribution obtained by MART

tion of the measurement apparatus as shown in Fig. 6. It was known that the peak values of the reconstructed temperature and H_2O concentration were 429.7K and 2.13%, respectively.

Conclusions

The MART (multiplicative algebraic reconstruction

technique) was newly adopted and its performances were validated for CT-TDLAS (computed tomography-tunable diode laser absorption spectroscopy).

It was validated that the calculation errors by MART algorithm were less than those by ART algorithm.

Further, the calculation speed of MART algorithm was 5 times faster than that of ART algorithm.

The minimum iteration number to be saturated at the minimum values of error was 10. This implies that several dozens of iteration number would be enough for attaining the minimum errors.

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